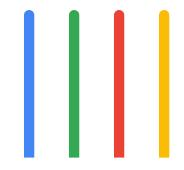


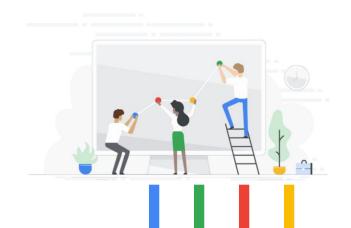
# เปิดประตูสู่โลกแห่งข้อมูลด้วย เครื่องมือด้าน Data บน

# Google Cloud











Boonrat (Lucky) Chanchoke



Plub Wittawin Waiyapat



## Google's Data Cloud

Simple. Limitless. Intelligent.



## **Google Mission Statement**



















Organize the world's information and make it universally accessible and useful





Google Search

I'm Feeling Lucky



### The innovation leader in applying Data and AI to real-world situations



#### Search

Search ranking Speech recognition



#### **Translate**

Text, graphic and speech translations



#### **Photos**

Photos search



#### **Gmail**

Smart reply Spam classification



Self Driving Car 20B miles driven



Smarter & Cleaner Infrastructure 2X more efficient



AlphaGo First Al to beat a world Go champion (2016)



#### YouTube

Video recommendations Better thumbnails



## Challenge #1

# Data is big and multi-format.

- Structured and unstructured
- Real-time streams and at-rest
- Across clouds and on-premise



- <u>STATISTA</u>, FEB 2022

## Challenge #2

# Data requires more than SQL.

- Machine learning & Al
- Stream analytics and events
- Data-driven applications



Gartner <sup>®</sup>, Streaming Analytics in the Cloud: A Comparative Analysis of Amazon, Microsoft and Google, Sumit Pal, Shaurya Rana, 14 December, 2021

## Challenge #3

# Data needs to reach everyone.

- Mission critical
- Accessed by everyone
- Governed



- <u>HBR</u>, 2021

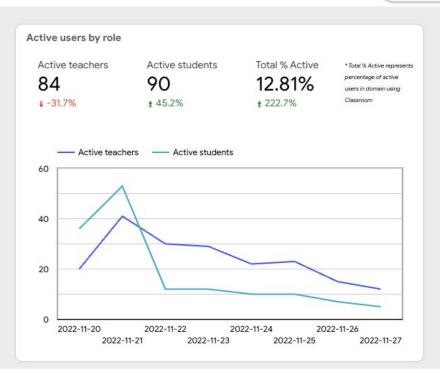


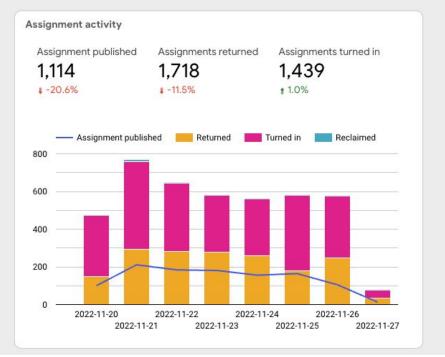
#### Classroom engagement report

Org unit

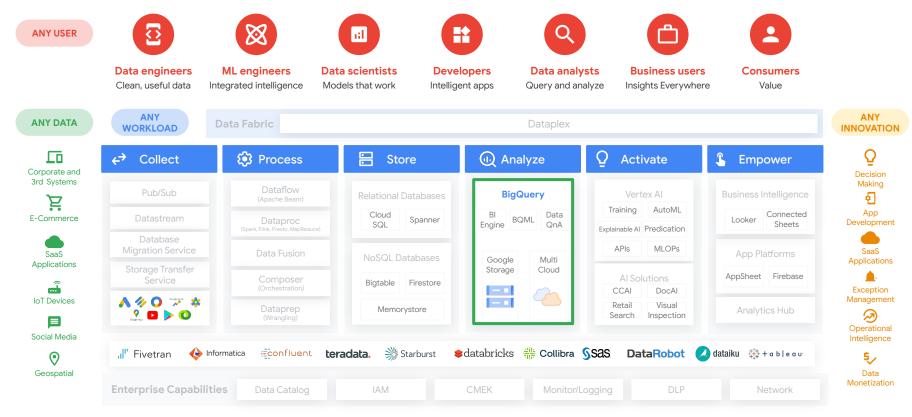
Last BQ Export Date 2022-11-27

Nov 20, 2022 - Nov 26, 2022





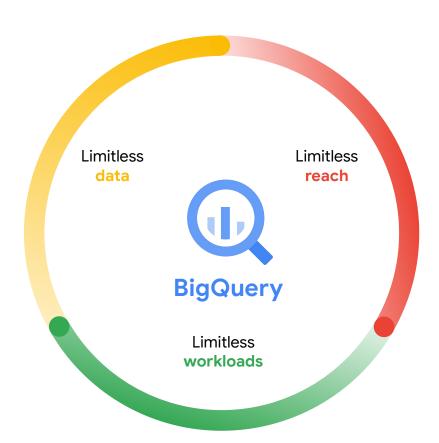
### A data platform that supports any needs, for any personas



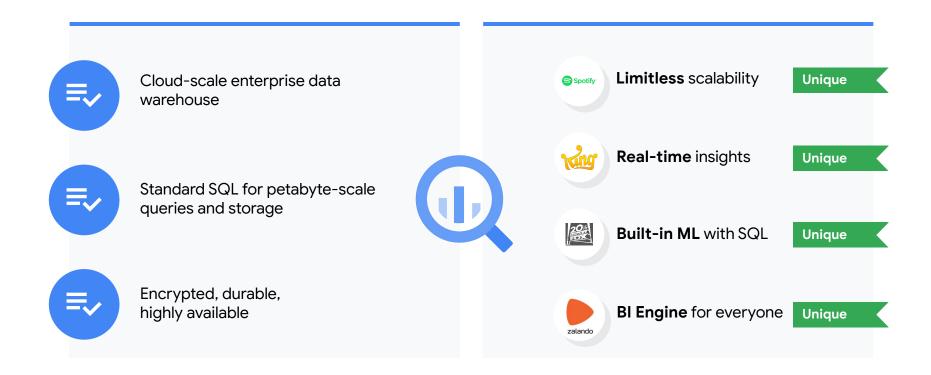
# **BigQuery**

The core of Google's Data Cloud to power your data-driven innovation.

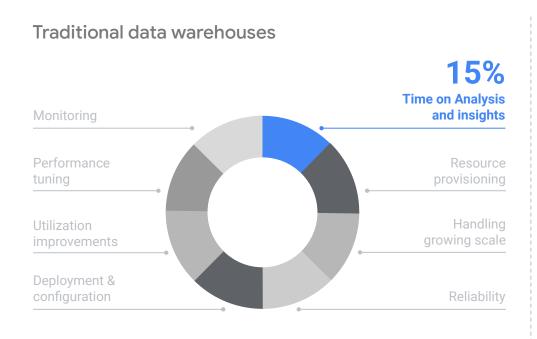
**100k+** data professionals have started their Data Cloud journey using BigQuery with trials growing nearly **150% YoY in 2021**.

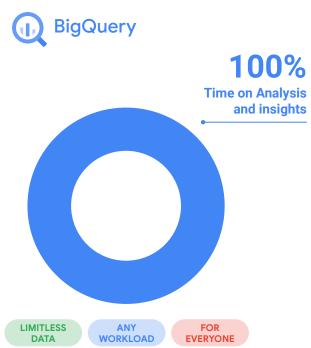


## Data warehouse with customers ranging from TB to 100+ PB



## Serverless data warehouse means you spend 100% using data

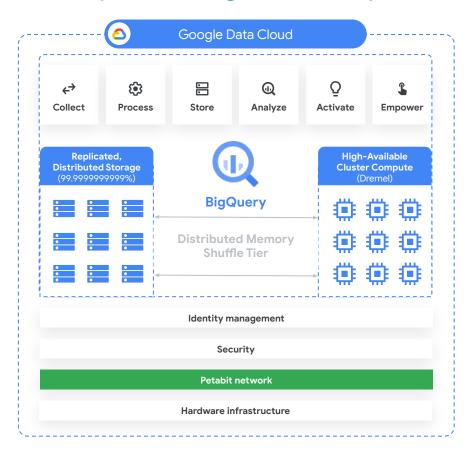




# **LIMITLESS DATA**

Limitless Data

## Decoupled storage and compute for maximum flexibility



#### **Completely elastic**

Distributed storage and compute with ultra-high bandwidth including distribute petabyte scale in-memory storage for temp data and state:

- Auto-start / auto-pause
- 0-Second warm up to get maximum performance
- No performance cliff due to local capacity saturation
- Immune to large-scale hardware failures

**Limitless Data** 

## BigQuery @ Spotify



BigQuery reduced the time for running common queries

# from minutes to seconds

#### **Question:**

How many hours did users in Spain spend listening to Spotify in October, 2016?



Hive 15.5TB processed



**BigQuery** 750GB processed



# Let's query 2TB dataset!

(57,000,000,000 rows)

Just load data, then query. No start up, no cluster.

# ANY WORKLOADS

Any Workloads

Proprietary + Confidential

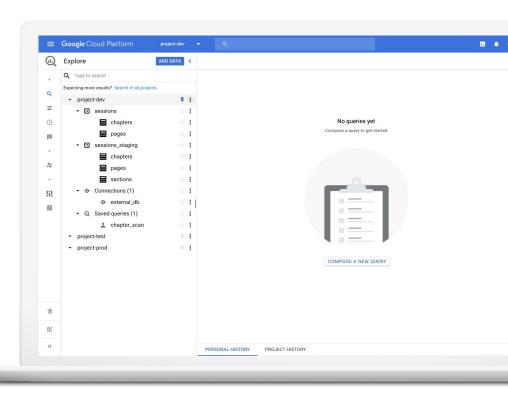
## Unified SQL and Spark experience

#### **Unified SQL and Spark experience**

Enable data warehousing users to easily write and execute Spark on BigQuery data **without** exporting it

#### Fully-managed Spark and SQL analytics

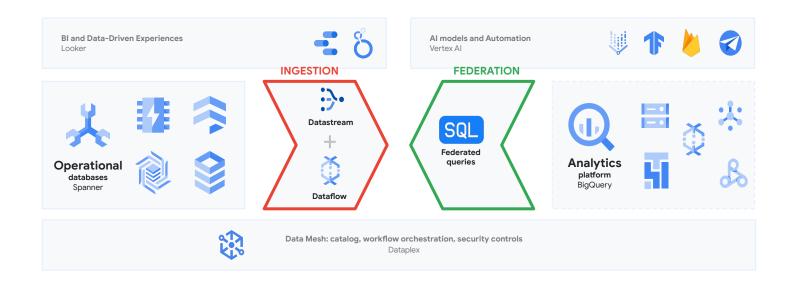
No infrastructure management required for either Spark or SQL analytics. **Both autoscale.** 



Any Workloads

## Analytical + Transactional OLAP and OLTP in one place

Analyze Cloud Spanner and Cloud SQL data in real-time without data movement or copy.



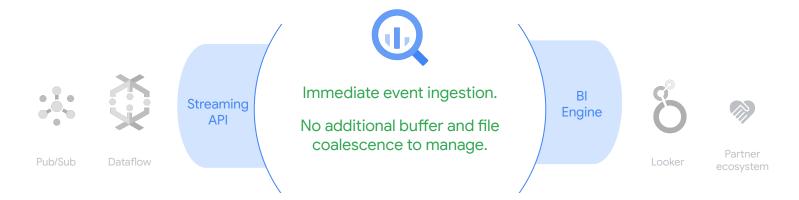
Any Workloads

Proprietary + Confidentia

## Stream Analytics right in the data warehouse

#### **Analyze business events in real-time**

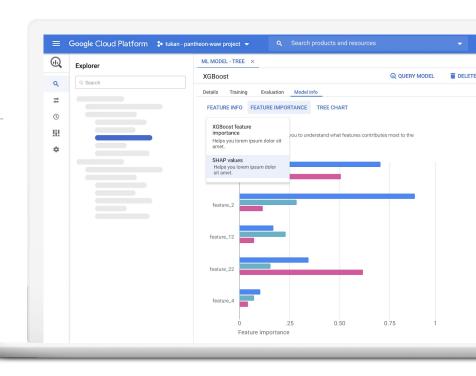
Move your business to event-driven action for logstream, clickstream, and sensor data to enable use cases like anomaly detection and continuous intelligence.

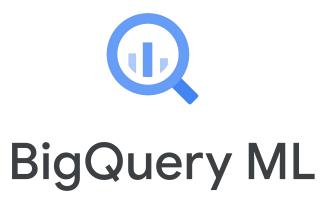


## BigQuery ML Built-in Al/ML + Explainable Al (XAI)

#### Machine Learning for all Built-in ML with SQL

- No Data Movement Execute, iterate, and automate ML initiatives all without data movement
- Pre-built algorithms plus leveraging external models developed in Tensorflow directly from SQL
- Production Ready Export developed models for production deployment in Vertex AI
- Complete MLOps environment, seamlessly integrated with Google Vertex AI





Now supports unstructured data



Paths

Google Cloud Skills Boost

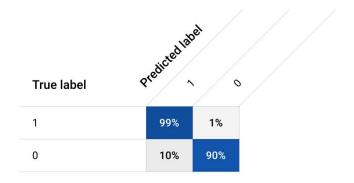
# Let's build some ML in BigQuery!

#### Score threshold

Positive class threshold	0	0162
Positive class	1	
Negative class	0	
Precision ?	0.0904	
Recall ?	0.9895	
Accuracy ?	0.9037	
F1 score ?	0.1656	

#### **Confusion matrix**

This table shows how often the model classified each label correctly (in blue),



# **FOR EVERYONE**

## A platform for all users and intents throughout the data life cycle



#### Developer

Intelligent apps



Messaging Pub/Sub



Data Processing Dataflow



OSS Engines Dataproc (Spark, Flink)



DW & DB BigQuery, Cloud Bigtable



Data Apps Looker (LookML)



#### Data engineer

Get clean, useful data



Messaging Pub/Sub or Confluent Kafka



Data processing (OSS) pipelines Dataproc (Spark, Presto, Flink)



Data Processing (native) pipelines Dataflow



Dataplex BigQuery & Cloud Storage



Orchestration Cloud Composer



#### Data analyst

Query and analyze



Ingestion BigQuery Streaming & DTS



Visual data Integration Cloud Data Fusion



Data warehouse & Orchestration BigQuery



Data models, catalog Looker, Data Catalog



ML in SQL BigQuery ML



#### **Business user**

Insights everywhere



Data warehouse BigQuery



Governed BI Looker



Data warehouse in a spreadsheet Connected Sheets



Data models, catalog Looker, Data Catalog



Natural Language Query



#### Data scientist

Models that work



#### Dataplex

BigQuery & Cloud Storage



Portable notebooks User-managed notebooks



Simplified ML BigQuery ML & AutoML



Spark Dataproc



Collaboration Feature Store, Vertex Al



#### Security admin

Protecting data



Fine-grained access control IAM



Metadata management Data Catalog



Redact sensitive data Cloud DLP

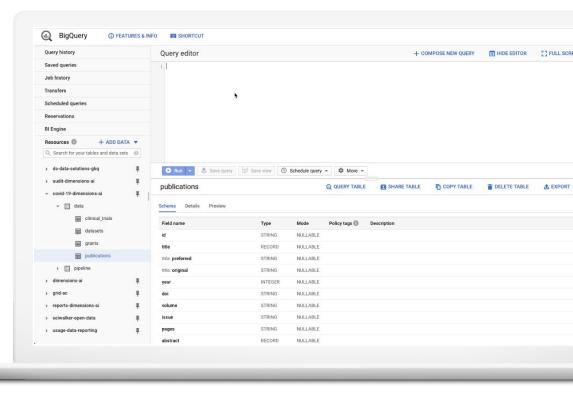


encrypted Data at rest and in transit



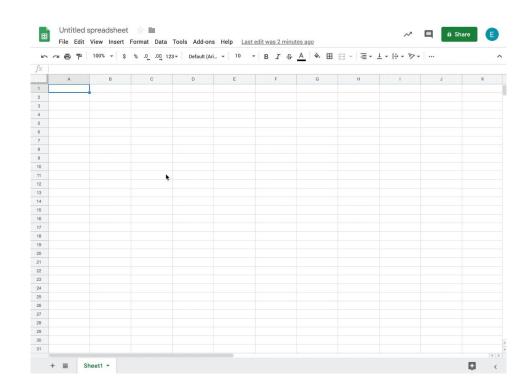
Secure data sharing BigQuery See your BigQuery data in one click with Looker Studio





Work with spreadsheet in Sheets, analyzing data with pivot table in a few clicks

Powered by BigQuery on the backend.





#### **Sheets**

Easy to use & shareable

Familiar interface

Light-weight analysis



#### Connected **Sheets**

Analyze billions of rows of data in Sheets. without any need for specialized knowledge



#### **BigQuery**

Analyze petabytes of data

Complex queries

Increase time to insight

Use data from Google BigQuery with the BigQuery data source / Connected Sheets in AppSheet app.

























Driving insights from your data with maps visualization in a few steps in BigQuery Geo Viz

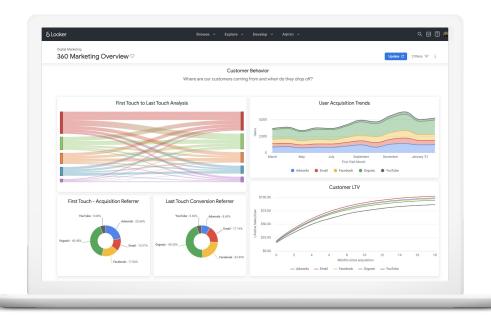
1) Select data in BigQuery

2) Select column In Geo Viz 3) Style the viz interactively

#### Looker

# The Platform for first-party data

Google's cloud-native enterprise data platform for accessing, analyzing, and activating first-party data.

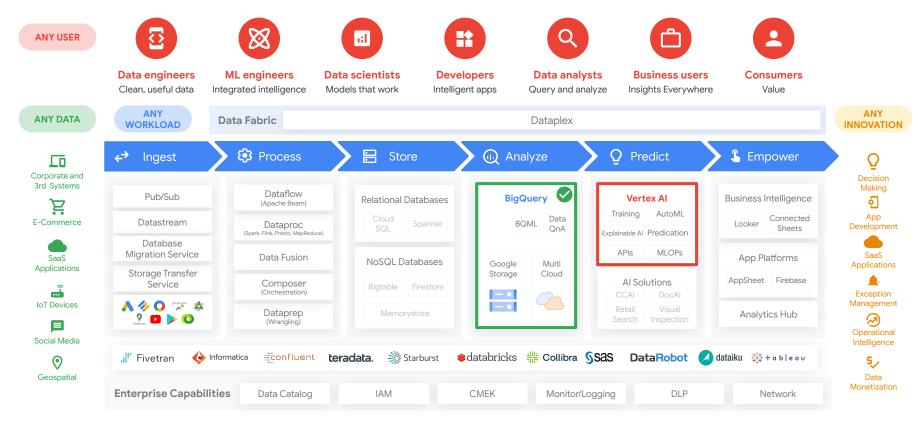


## BigQuery helps you do more

Limitless data	All workloads	For everyone
Completely serverless  All data types	SQL, Spark, Search, Stream  Built-in Al/ML	Built-in BI Analytical applications
Data exchanges	Analytical + Transactional	Analytical applications  Partner ecosystem

Cost Effective | Highly Productive | Governed | Easy to Secure | Clear Compliance | Open Extensions

## A data platform that supports any needs, for any personas



### Vertex AI has multiple pathways of consumption

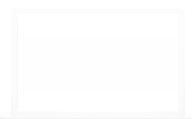
offering a unified platform to build, deploy and scale ML models with innovations developed by Google

Out of the box





- Pre-trained APIs & Al Solutions
  - Call API to predict
  - Pre-trained
  - No training data needed
  - Get started right away



- 2 BigQuery ML
- Simple SQL code
- Descriptive and predictive modeling on structured data
- Hyper-parameter tuning
- Feature engineering
- Explainability



- 3 AutoML in Vertex A
- No code, form-driven
- Structured & unstructured data
- Hyper-parameter tuning
- Feature engineering
- Explainability

- 4 End-to-end MLOps
- Code on managed infra
- Custom models on pre-built frameworks
- Noops, serverless training with hyperparameter tuning
- Explainability



# 1 Pre-trained Al APIs

### Sight



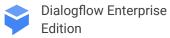


### Language



Natural Language API

### Conversation







### **Al Solutions**



Document Al





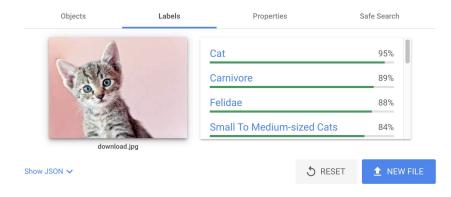
# 1

### **Pre-trained Al APIs: Vision API**

```
import io
import os
# Imports the Google Cloud client library
from google.cloud import vision
# Instantiates a client
client = vision.ImageAnnotatorClient()
# The name of the image file to annotate
file_name = os.path.abspath('resources/wakeupcat.jpg')
# Loads the image into memory
with io.open(file_name, 'rb') as image_file:
   content = image_file.read()
image = vision.Image(content=content)
# Performs label detection on the image file
response = client.label_detection(image=image)
labels = response.label_annotations
print('Labels:')
for label in labels:
   print(label.description)
```

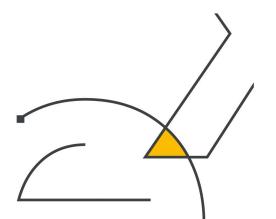
### https://cloud.google.com/vision/docs/drag-and-drop

#### Try the API



### What is Document AI?

Document Al turns unstructured content into business-ready structured data.



### Document image



### Structured data

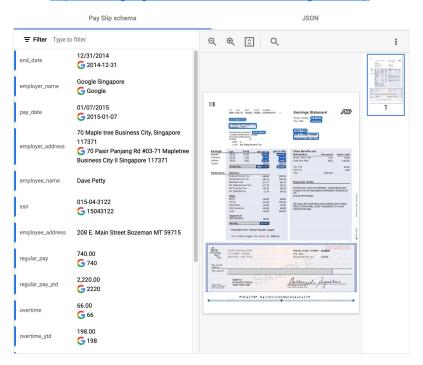
```
{Class:
/us/gov/ID/Driverslicense
    State: Ohio
    Name: Jane Doe
    Issued: 7/2/2018
    Expires: 1/21/2021
}
```

# 1

### Al Solutions: Document Al

```
from google.cloud import documentai_v1 as documentai
# You must set the api_endpoint if you use a location other than 'us', e.g.:
opts = \{\}
if location == "eu":
    opts = {"api_endpoint": "eu-documentai.googleapis.com"}
client = documentai.DocumentProcessorServiceClient(client_options=opts)
# The full resource name of the processor, e.g.:
# projects/project-id/locations/location/processor/processor-id
# You must create new processors in the Cloud Console first
name = f"projects/{project_id}/locations/{location}/processors/{processor_id}
# Read the file into memory
with open(file_path, "rb") as image:
    image_content = image.read()
document = {"content": image_content, "mime_type": "application/pdf"}
# Configure the process request
request = {"name": name, "raw_document": document}
result = client.process_document(request=request)
document = result document
document_pages = document.pages
# For a full list of Document object attributes, please reference this page: h
# Read the text recognition output from the processor
print("The document contains the following paragraphs:")
for page in document_pages:
    paragraphs = page.paragraphs
    for paragraph in paragraphs:
        print(paragraph)
        paragraph_text = get_text(paragraph.layout, document)
        print(f"Paragraph text: {paragraph_text}")
```

#### https://cloud.google.com/document-ai/docs/drag-and-drop



# **Document AI Processors**



### General

Your Content Google's General Models



### **Specialized**

Your Content Google's Specialized Models



### **Document Al Workbench**

Automate document processing by using your data to build models with Document Al's machine learning platform powered by state of the art computer vision, NLP, and neural networks.



# Why Document Al Workbench?

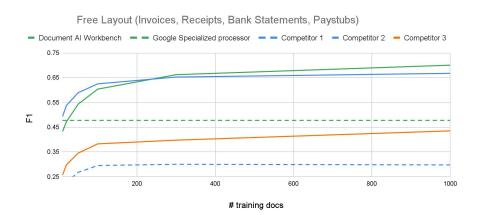
1 Any doc type	Use your data to create ML models with DocAl Workbench for many doc types (e.g., printed, scanned, handwritten, tables, etc.)
2 Democratized ML	Label data with a simple interface; train ML models with the click of a button
3 Time to market (TTM)	Reduce TTM with Workbench instead of building custom ML models
4 Less training data	Transfer learnings (aka "uptrain") from relevant models to get accurate results faster for your documents
5 No cost training	Create and evaluate ML models for free; pay as you go once you deploy and use processors to extract document data
6 Own your data	You own your data within your GCP project



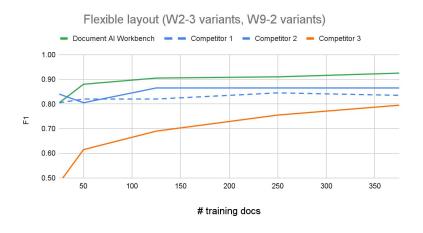
### How do we stack up with competitors?

**Context:** A 3-party agency used Document Al Workbench alongside major competitors' products to automate document processing

**Note**: The below F1 scores (harmonic mean of precision & recall) estimate model prediction accuracy. Higher accuracy (99+%) can be achieved through human in the loop (HITL) review.



What we learned: Workbench extracted data from documents more accurately and with less training data for flexible (W9-2 variants, W2-3 variants) and free layout (invoice, receipt, bank statements, paystubs) document types



### Vertex AI has multiple pathways of consumption

offering a unified platform to build, deploy and scale ML models with innovations developed by Google

Out of the box





- 1 Pre-trained APIs & AI Solutions
  - Call API to predict
  - Pre-trained
  - No training data needed
  - Get started right away



- 2 BigQuery ML
- Simple SQL code
- Descriptive and predictive modeling on structured data
- Hyper-parameter tuning
- Feature engineering
- Explainability



- 3 AutoML in Vertex Al
- No code, form-driven
- Structured & unstructured data
- Hyper-parameter tuning
- Feature engineering
- Explainability



- 4 End-to-end MLOps
- Code on managed infra
- Custom models on pre-built frameworks
- Noops, serverless training with hyperparameter tuning
- Explainability

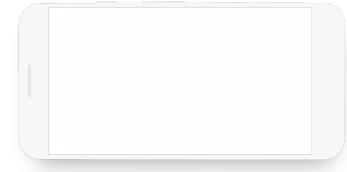


### Machine Learning for all Built-in ML with SQL

- Execute, iterate, and automate ML initiatives all within BigQuery using predefined models
- Leverage external models developed in Tensorflow directly from SQL
- Export developed models for use in Vertex Al

### **BigQuery ML**

Train models in SQL; manage, orchestrate, and deploy directly to Vertex Al

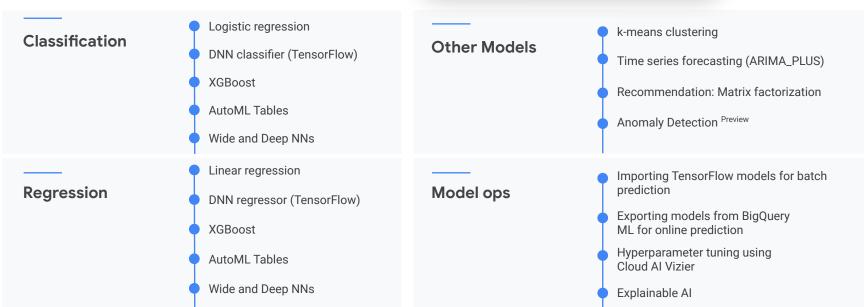














### Vertex AI has multiple pathways of consumption

offering a unified platform to build, deploy and scale ML models with innovations developed by Google

Out of the box





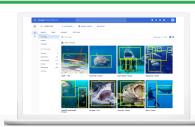
**Pre-trained APIs & AI Solutions** 



2 BigQuery ML

- Call API to predict
- Pre-trained
- No training data needed
- Get started right away

- Simple SQL code
- Descriptive and predictive modeling on structured data
- Hyper-parameter tuning
- Feature engineering
- Explainability



- 3 AutoML in Vertex Al
- No code, form-driven
- Structured & unstructured data
- Hyper-parameter tuning
- Feature engineering
- Explainability

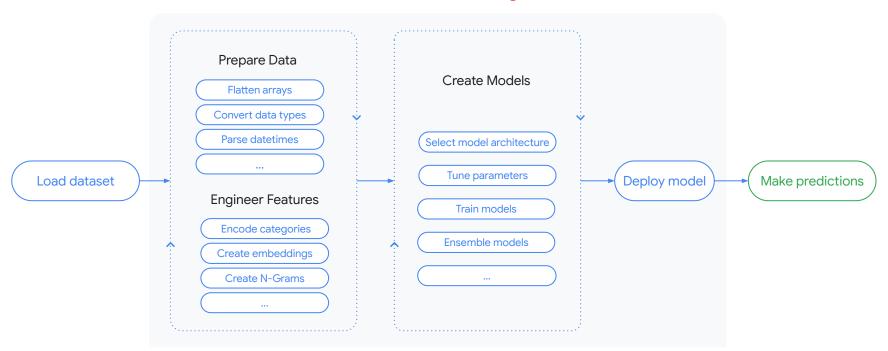


- 4 End-to-end MLOps
- Code on managed infra
- Custom models on pre-built frameworks
- Noops, serverless training with hyperparameter tuning
- Explainability



### AutoML - Fastest path from data to value

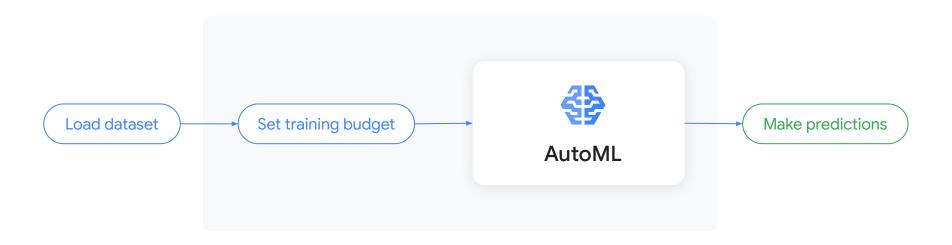
### **Traditional Machine Learning Workflow**



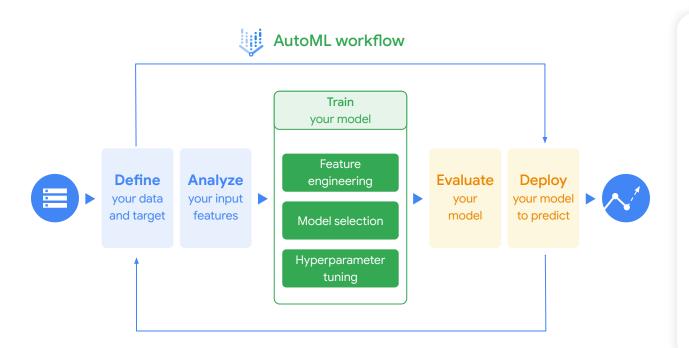
# 3

### AutoML - Fastest path from data to value

### **AutoML Workflow**



### No Code: Point and click to build custom, high-quality models



Automatically search through Google's whole model zoo...

Linear, logistic

Feedforward DNN

Wide and Deep NN

Gradient Boosted Decision Tree (GBDT)

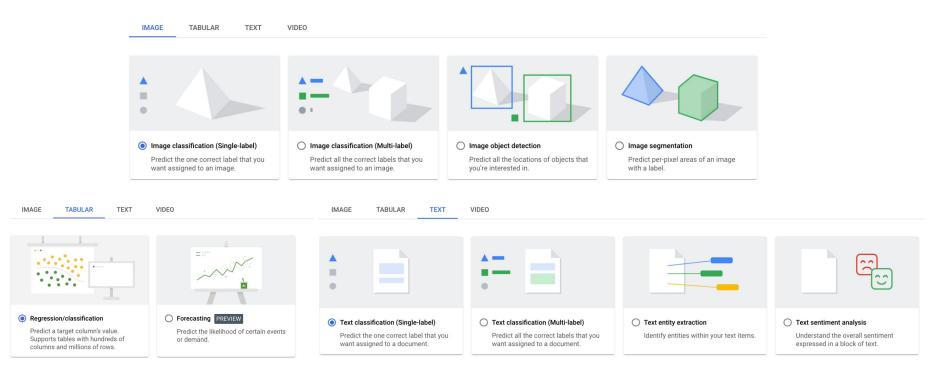
DNN + GBDT Hybrid

Adanet ensemble

Neural + Tree Architecture Search

...and more!

### 3 AutoML - Various types of models supported



# 3 AutoML - Case Study



98%

precision with AutoML compared to 70-80% with custom models

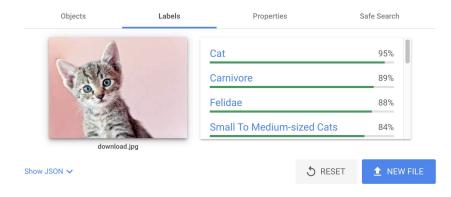
# 1

### **Pre-trained Al APIs: Vision API**

```
import io
import os
# Imports the Google Cloud client library
from google.cloud import vision
# Instantiates a client
client = vision.ImageAnnotatorClient()
# The name of the image file to annotate
file_name = os.path.abspath('resources/wakeupcat.jpg')
# Loads the image into memory
with io.open(file_name, 'rb') as image_file:
   content = image_file.read()
image = vision.Image(content=content)
# Performs label detection on the image file
response = client.label_detection(image=image)
labels = response.label_annotations
print('Labels:')
for label in labels:
   print(label.description)
```

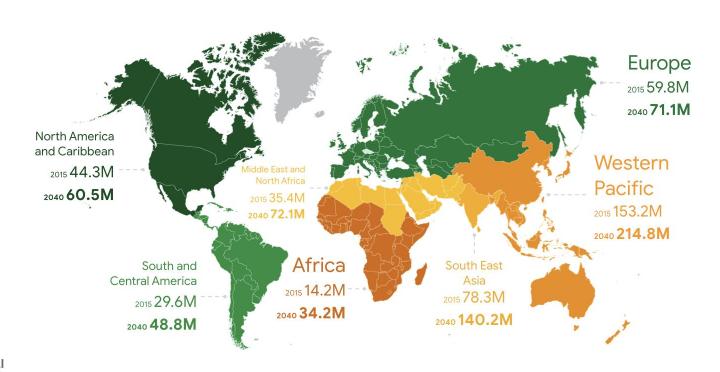
#### https://cloud.google.com/vision/docs/drag-and-drop

#### Try the API



# Diabetic retinopathy: fastest growing cause of blindness

### **415M** people with diabetes

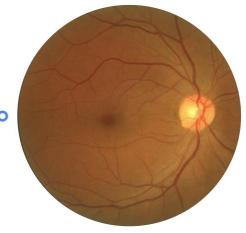






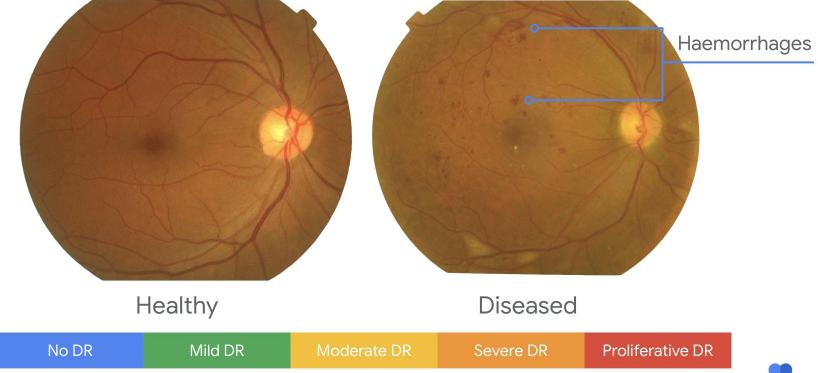
# Regular screening is key to preventing blindness







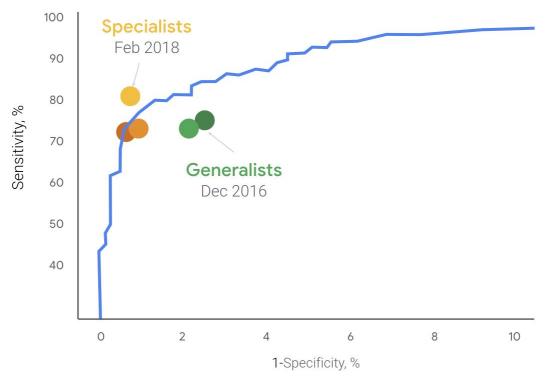
# How DR is Dianosed: Retinal Fundus Images





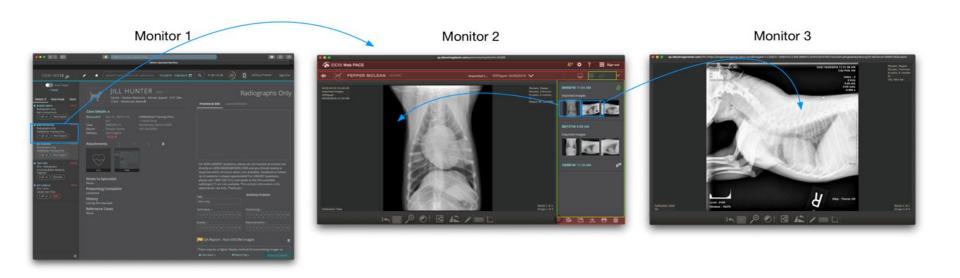


# Al's performance is on par with eye specialists





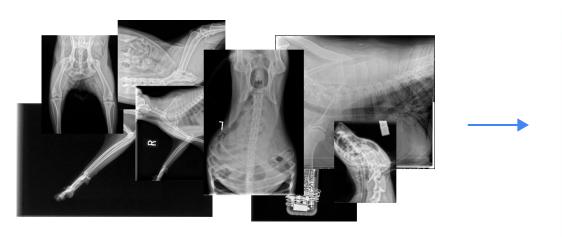
### **VetMedStat Telemedicine workflow**







## Challenge: Images come in unorganized



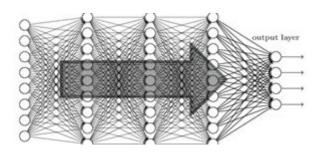






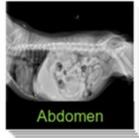
## **Use Machine Learning to Organize Data**









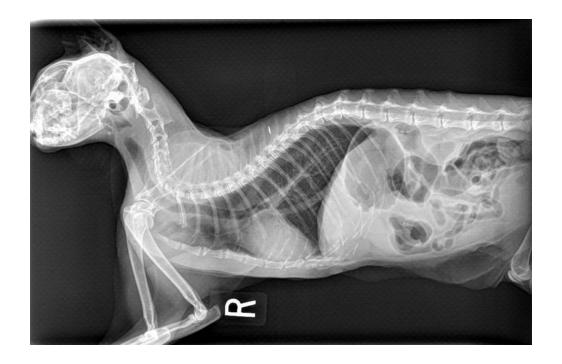








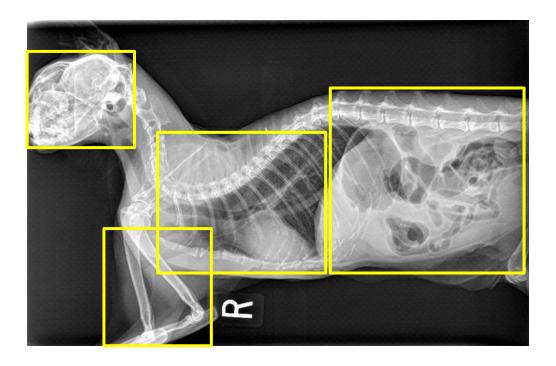
# What kind of image is this? Skull? Thorax?







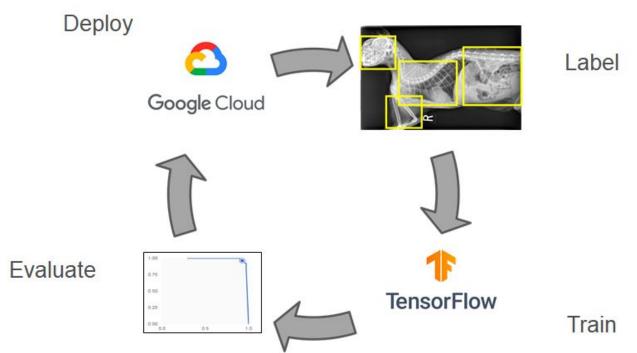
# Object Detection gives us flexibility







# Typical ML project cycle is long









# **Vertex Al Vision**

Reduce time to build computer vision applications from days to minutes at one tenth the cost of current offerings

Public preview

# An explosion of connected cameras and Al is transforming industries







### Enterprise

- Retail
- Financial services
- Manufacturing
- Transportation

- Food & beverages
- Healthcare & hospitality
- Construction
- Education

### Public sector

- Smart Cities
- Traffic management Systems
- Government Infrastructures

### Consumer

- Smart Homes
- Fitness
- Media, entertainment
- Gaming

# Building Computer Vision apps isn't easy.

### **Computer Vision Today**



Complex



**Expensive** 



**Untrusted** 

### **Vertex Al Vision**



Easy to use



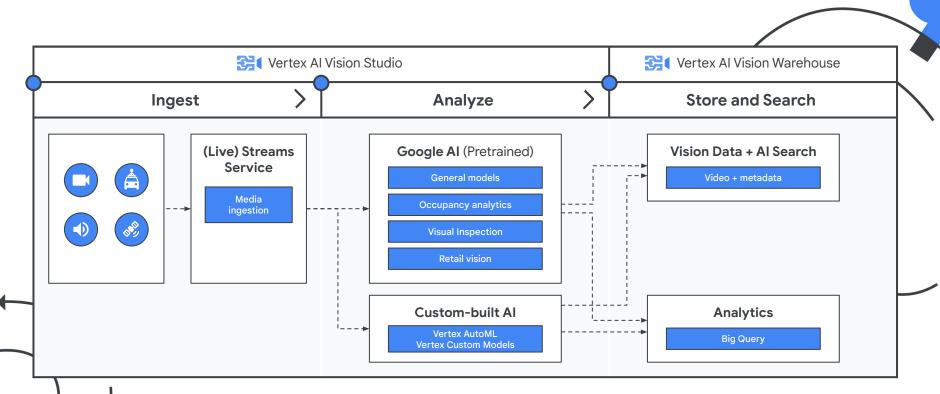
Low cost



Responsible Al

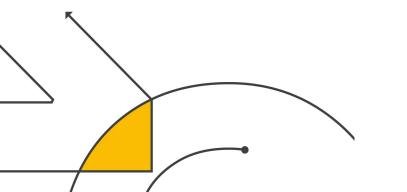
### What is Vertex Al Vision?

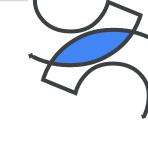
One Stop shop - Simple interface, serverless and scalable platform to quickly build and deploy Vision AI applications



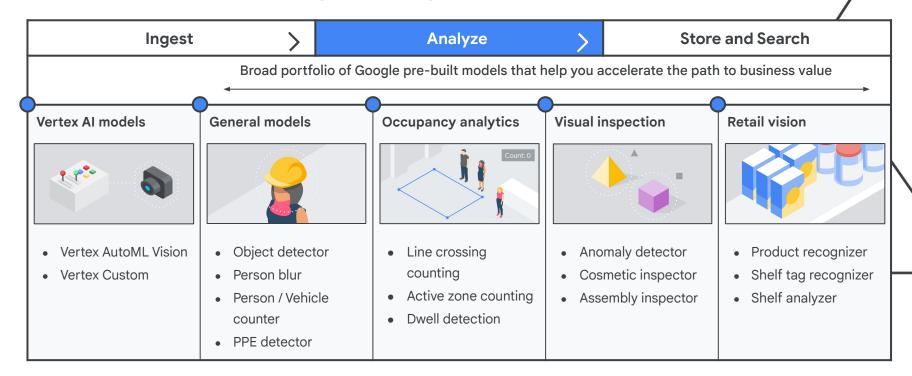


# Demo



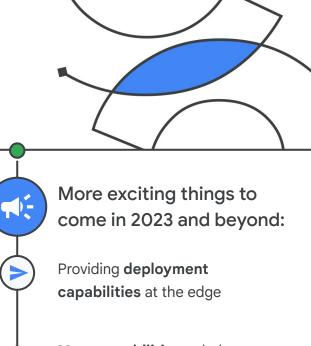


# Vertex Al Vision makes it easy to build your own Al models or leverage Google prebuilt models



#### Looking forward & conclusion

Vertex Al Vision: Next generation business insights





Vertex Al Vision brings:



Partners complement **Vertex Al Vision:** 





No code / low code that lets more users build Vision Al apps for your business



Partners bring key capabilities and domain experience with Vertex Al Vision



SOTA Al that is cost efficient



Partners integrate our Vertex Al Vision platform and models into their solutions



More capabilities to help you build your own models

#### Vertex AI has multiple pathways of consumption

offering a unified platform to build, deploy and scale ML models with innovations developed by Google

Out of the box











- 1 Pre-trained APIs & AI Solutions
  - Call API to predict
  - Pre-trained
  - No training data needed
  - Get started right away

- 2 BigQuery ML
- Simple SQL code
- Descriptive and predictive modeling on structured data
- Hyper-parameter tuning
- Feature engineering
- Explainability

- 3 AutoML in Vertex Al
- No code, form-driven
- Structured & unstructured data
- Hyper-parameter tuning
- Feature engineering
- Explainability

- 4 End-to-end MLOps
- Code on managed infra
- Custom models on pre-built frameworks
- Noops, serverless training with hyperparameter tuning
- Explainability





## **Google Data Studio**

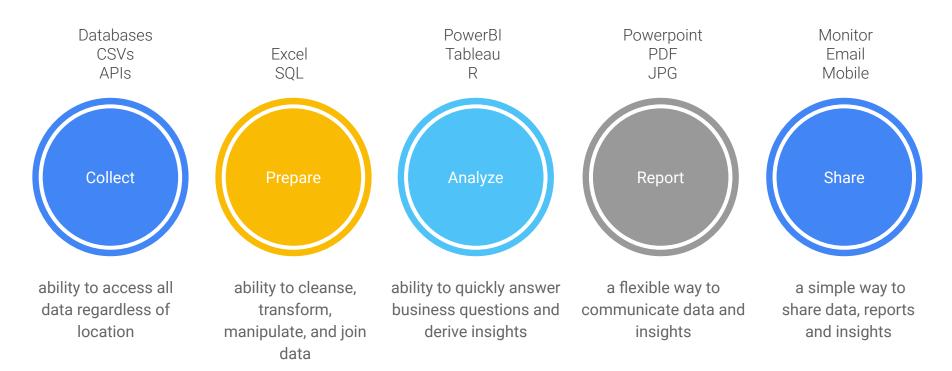
Hands-on training



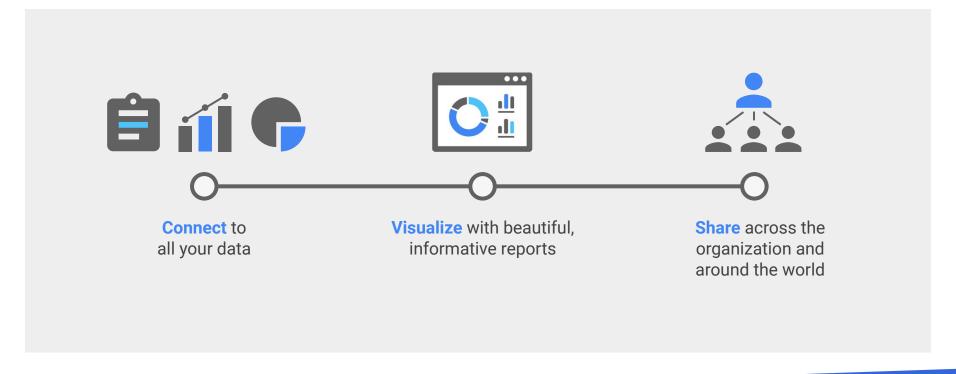


# About Looker Studio

#### All businesses manage data the same way



# Looker Studio: Google's BI/Reporting tool available externally



### Why DLookerata Studio?

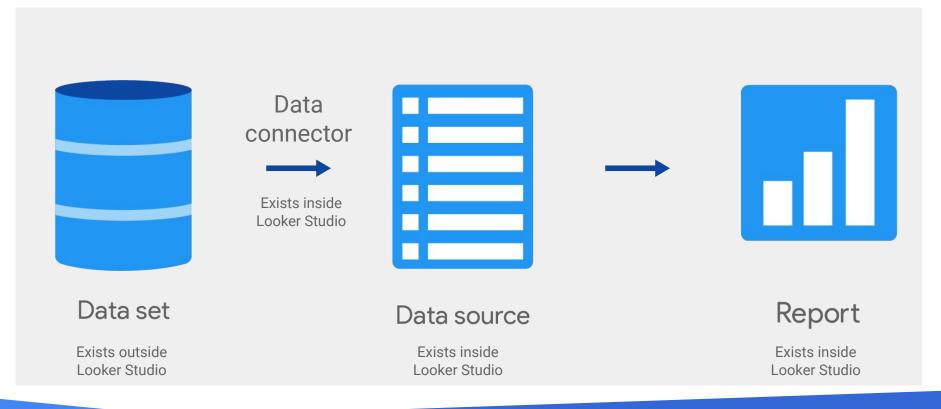
Looker Studio provides tools to create **beautiful reports** & perform **powerful ad-hoc analysis**.

- Large number of data connectors
- Easy to maintain
- Easy to use
- Tech skills not mandatory
- Easy sharing and collaboration
- Free & globally available



# Looker Studio building blocks

## Data entity relationship



### Data sources connect to underlying data sets

2 basic types of data sets:

#### **Fixed schema**

We understand the data before we ingest it

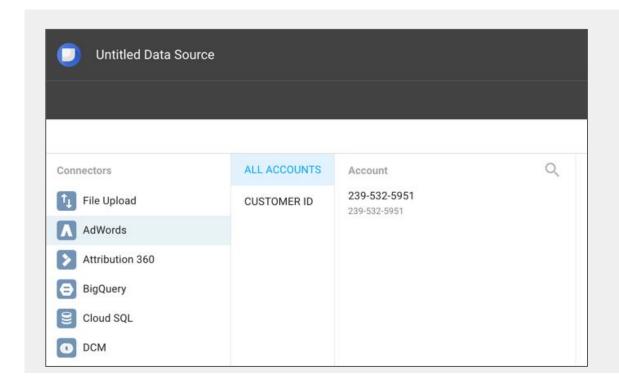
- Google Ads
- Search Ads 360
- Google Analytics
- Search Console
- YouTube Analytics
- Display & Video 360

#### Flexible schema

No idea what the data is before we ingest it

- BigQuery
- File upload (CSV)
- Google Cloud Storage
- Google Sheets
- SQL connectors (MySQL, PostgreSQL)

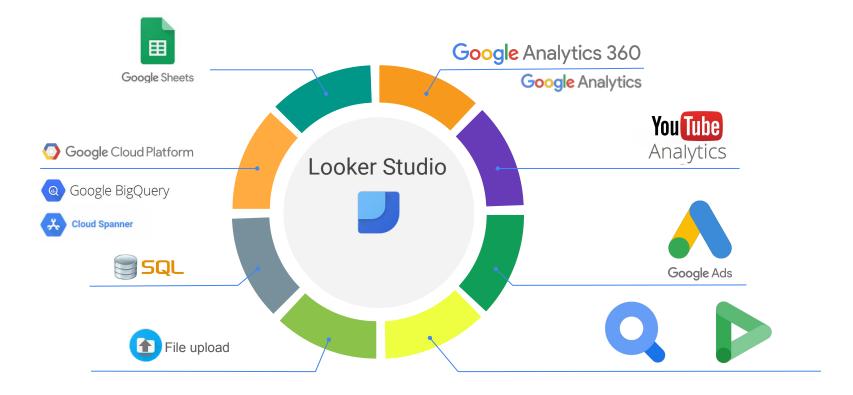
#### Connectors are pipes to the data set



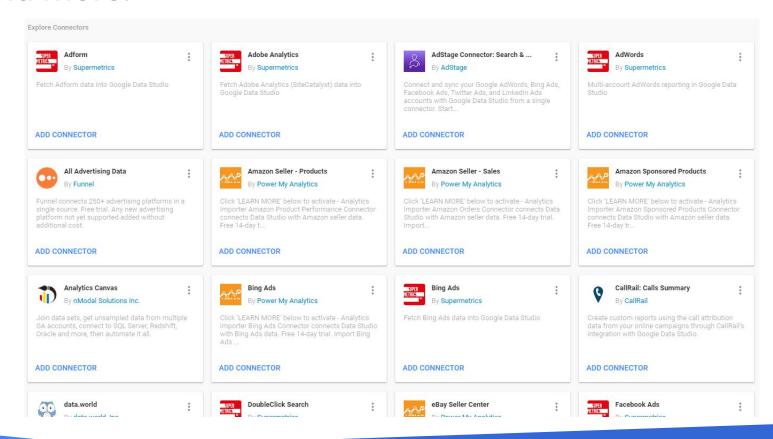
#### Connecting is easy-peasy!

- Select the connector
- Select the account, file, project, etc.
- Provide credentials
- Access via OAuth

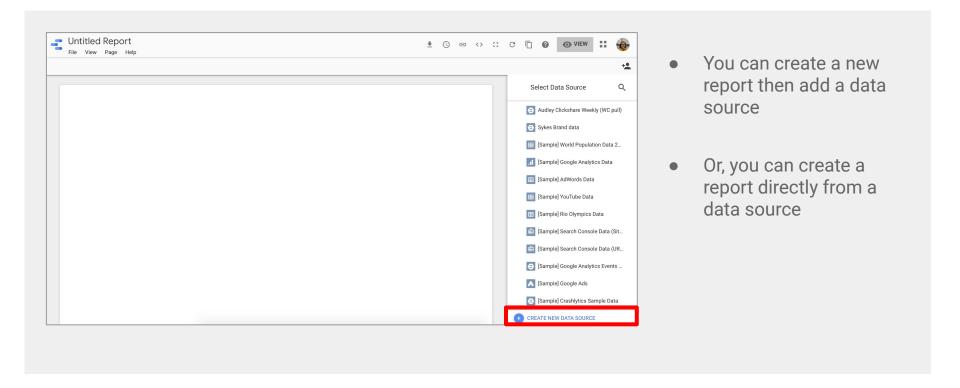
#### Looker Studio connectors



#### ... and more!



### Reports are built from data sources





# Looker Studio home page

lookerstudio.google.com